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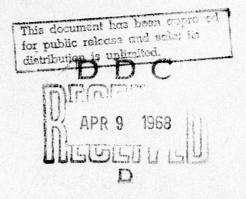
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Generating Associated Random Variables

James D. Esary Frank Proschan



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Boeing Scientific Research Laboratories Seattle, Washington

January 1968

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GENERATING ASSOCIATED RANDOM VARIABLES

bу

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and

Frank Proschan

Mathematical Note No. 546

Mathematics Research Laboratory

BOEING SCIENTIFIC RESEARCH LABORATORIES

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Abstract

Random variables $T = \{T_1, \ldots, T_n\}$ are associated if $Cov[f(\underline{T}), g(\underline{T})] \geq 0$ for all increasing f,g for which the covariance exists. T_{n+1} is stochastically increasing in T_1, \ldots, T_n if $P[T_{n+1} > t_{n+1} \mid T_1 = t_1, \ldots, T_n = t_n]$ is increasing in t_1, \ldots, t_n for each fixed t_{n+1} . In this paper, results of the following type are derived: If T_i is stochastically increasing in T_1, \ldots, T_{i-1} for $i = 1, \ldots, n$, then T_1, \ldots, T_n are associated. Examples are given of the application of these results to reliability models involving various types of maintenance.

Acknowledgement

We would like to thank A. W. Marshall and R. Pyke for their helpful suggestions and advice.

Generating Associated Random Variables

1. <u>Introduction</u>

A set of random variables $\underline{T} = \{T_1, \dots, T_n\}$ are said to be associated, if $Cov[f(\underline{T}), g(\underline{T})] \geq 0$ for all increasing functions f,g for which the covariance exists (an increasing function is a function which is nondecreasing in each of its arguments). Esary-Proschan-Walkup (hereafter referred to as E-P-W) (1967) develop the basic properties of associated random variables and present some applications [see also E-P-W (1966) for applications to reliability theory]. Tukey (1958) discusses the notion of positive regression dependence of T_2 on T_1 , defined by the property that $P[T_2 > t_2 \mid T_1 = t_1]$ is increasing in t_1 for each fixed t_2 . Lehmann (1966) discusses several forms of bivariate dependence, including positive regression dependence (but not bivariate association), shows their relationships, and gives a number of applications. Esary and Proschan (1967) discuss the relationships between bivariate association and the forms of bivariate dependence considered by Lehmann.

It is shown in E-P-W (1967) that positive regression dependence of T_2 on T_1 implies association between T_1 and T_2 . In the present paper we define T_{n+1} to be at chartifully increasing in T_1, \ldots, T_n if $P[T_{n+1}, t_{n+1} | T_1 = t_1, \ldots, T_n = t_n]$ is increasing in t_1, \ldots, t_n for each fixed t_{n+1} , and show:

Theorem 1.1. Let T_1, \ldots, T_n be associated. Let T_{n+1} be stochastically increasing in T_1, \ldots, T_n . Then T_1, \ldots, T_{n+1} are associated.

We say "stochastically increasing" rather than the previously introduced "positive regression dependent" in order to have a terminology consistent with the usual notion of stochastic ordering, which we find it convenient to employ.

Lehmann (1967), Example 1, considers the construction $S_1 = h_1(U_1,T)$, $S_2 = h_2(U_2,T)$, where U_1 , U_2 , T are independent and h_1 , h_2 are functions increasing in T. We show:

Theorem 1.2. Let $T_1, ..., T_n$ be associated. Let $S_i = h_i(U_i, T_1, ..., T_n)$, i = 1, ..., m, where $U_1, ..., U_m$ are mutually independent and also independent of $T_1, ..., T_n$ and h_i is increasing in $T_1, ..., T_n$. Then $S_1, ..., S_m$ are associated.

To prove Theorems 1.1 and 1.2 we consider a more general result (Theorem 3.1, or alternately Theorem 3.4) which includes both as special cases.

This investigation is primarily motivated by an implication of Theorem 1.1; random variables T_1, \ldots, T_n are associated if each T_i is stochastically increasing in T_1, \ldots, T_{i-1} . This fact is useful in reliability analyses involving maintenance, spares, and queueing for repair. See Section 4 for examples. We will discuss these applications in more detail in a forthcoming document on maintenance models.

2. Representation of Stochastically Increasing Random Variables

Let S and T be random variables. Let $S = \{S_1, \ldots, S_n\}$ and $T = \{T_1, \ldots, T_n\}$ be sets of random variables. S is stochastically equal to T, written $S = {}^{st}T$, if S and T have the same probability distribution. S is stochastically less than T, written $S \leq {}^{st}T$, if $P[S>u] \leq P[T>u]$ for all u. S is stochastically increasing in T, written $S \uparrow st$ in T, if $P[S>u]T=t^{(1)}] \leq P[S>u]T=t^{(2)}$ for all t.

Let $S \mid T = t$ denote a set of random variables with the conditional probability distribution of S, given that T = t.

We will use the following readily verified facts without further reference. $S,T=\overset{st}{S},T$ is equivalent to $S|T=t=\overset{st}{S},T=t$ for all t. $S \uparrow st$ in T is equivalent to $S|T=t^{(1)} \leq \overset{st}{S}|T=t^{(2)}$ for all $t^{(1)} \leq t^{(2)}$. $f(S,T)|T=t=t=\overset{st}{S}f(S,t)|T=t$, for any function f(S,t).

The following lemma is a variation on a basic result due to Lehmann (1959), p. 73.

Lemma 2.1. Let S + st in T. Then there exists an increasing function h(u,t), such that S,T = st h(U,T),T, where U is a random variable independent of T.

<u>Proof.</u> Let $F_{\underline{t}}$ be the distribution function of $S|\underline{T}=\underline{t}$, i.e., $F_{\underline{t}}(s) = P[S \le |\underline{T}=\underline{t}]$. Let $h(u,\underline{t}) = \inf\{s: u \le F_{\underline{t}}(s)\}$. h is increasing in u by its definition. S + st in \underline{T} implies $h(u,\underline{t}^{(1)}) \le h(u,\underline{t}^{(2)})$ for $\underline{t}^{(1)} \le \underline{t}^{(2)}$. Thus h is increasing in \underline{t} . Since $F_{\underline{t}}$ is

continuous from the right in s, $h(u,\underline{t}) \le s \iff u \le F_{\underline{t}}(s)$. Let U be uniformly distributed on [0,1]. Then $P[h(U,\underline{t}) \le s] = P[U \le F_{\underline{t}}(s)] = F_{\underline{t}}(s)$, i.e., $h(U,\underline{t}) = {}^{st}S|\underline{T} = \underline{t}$. Let U be independent of T. Then $h(U,\underline{t})|\underline{T} = \underline{t} = {}^{st}h(U,\underline{t})$. Thus

$$S \mid \underline{T} = \underline{t} = {}^{st} h(U,\underline{t}) = {}^{st} h(U,\underline{t}) \mid \underline{T} = \underline{t}$$

$$= {}^{st} h(U,\underline{T}) \mid \underline{T} = \underline{t}.$$

It follows that $S, \underline{T} = {}^{st} h(U,\underline{T}), \underline{T}. ||$

It is immediate that if $S = {}^{st} h(U,\underline{T})$, where $h(u,\underline{t})$ is increasing in \underline{t} and U is independent of \underline{T} , then $S \uparrow st$ in \underline{T} .

 S_1, \dots, S_m are conditionally independent, given that $\underline{T} = \underline{t}$, if $\underline{S} | \underline{T} = \underline{t} = {}^{st} \{ S_1 | \underline{T} = \underline{t}, \dots, S_m | \underline{T} = \underline{t} \}$, where $S_1 | \underline{T} = \underline{t}, \dots, S_m | \underline{T} = \underline{t}$ are assumed to be mutually independent.

Corollary 2.2. Let S_1, \ldots, S_m be conditionally independent, given $\underline{T} = \underline{t}$, for all \underline{t} . Let S_i st in \underline{T} , $i = 1, \ldots, m$. Then there exist increasing functions h_1, \ldots, h_m and mutually independent random variables U_1, \ldots, U_m that are independent of \underline{T} , such that

$$S,T = st\{h_1(U,I),...,h_m(U_m,I)\}, T.$$

<u>Proof.</u> Since S_i † st in T_i , set S_i , T_i = S_i h_i(U_i , T_i), T_i in accordance with Lemma 2.1. Then S_i | T_i = t_i = S_i h_i(U_i , t_i). Let U_i ,..., U_m be mutually independent. Then since S_i ,..., S_m are conditionally independent, given T_i = t_i ,

$$\begin{split} & \underset{\boldsymbol{\Sigma}}{\boldsymbol{\Sigma}} | \, \underline{\boldsymbol{T}} = \, \underline{\boldsymbol{t}} = \, \overset{\text{st}}{\boldsymbol{t}} \big\{ \boldsymbol{h}_{1}(\boldsymbol{U}_{1}, \underline{\boldsymbol{t}}) \,, \dots, \boldsymbol{h}_{m}(\boldsymbol{U}_{m}, \underline{\boldsymbol{t}}) \big\} \\ & = \, \overset{\text{st}}{\boldsymbol{t}} \big\{ \boldsymbol{h}_{1}(\boldsymbol{U}_{1}, \underline{\boldsymbol{t}}) \,, \dots, \boldsymbol{h}_{m}(\boldsymbol{U}_{m}, \underline{\boldsymbol{t}}) \big\} \big| \, \underline{\boldsymbol{T}} = \, \underline{\boldsymbol{t}} \\ & = \, \overset{\text{st}}{\boldsymbol{t}} \big\{ \boldsymbol{h}_{1}(\boldsymbol{U}_{1}, \underline{\boldsymbol{T}}) \,, \dots, \boldsymbol{h}_{m}(\boldsymbol{U}_{m}, \underline{\boldsymbol{T}}) \big\} \big| \, \underline{\boldsymbol{T}} = \, \underline{\boldsymbol{t}}. \end{split}$$

Thus $S, T = {}^{st}\{h_1(U_1,T),\dots,h_m(U_m,T),T,\dots,T,\dots\}$

Theorem 2.3. Let S_1, \dots, S_m be conditionally independent, given $T \cdot t$, for all t. Let $S_i \uparrow st$ in T, $i = 1, \dots, m$. Let $f(\underline{s}, \underline{t})$ be an increasing function. Then $f(\underline{s}, \underline{t}) \uparrow st$ in T.

Proof. Set $S,T = {st \{h_1(U_1,T), \ldots, h_m(U_m,T)\}, T \text{ in accordance with } Corollary 2.2.}$ Let $\xi(\underline{u},\underline{t}) = f[h_1(\underline{u},\underline{t}), \ldots, h_m(\underline{u},\underline{t}),\underline{t}].$ Then ξ is increasing, and $f(S,T) = {st \{(\underline{U},T).}$ For $\underline{t}^{(1)} \leq \underline{t}^{(2)}$

$$\xi(\underline{\mathbb{U}},\underline{\mathbb{T}})|\underline{\mathbb{T}} = \underline{\mathbb{t}}^{(1)} = \mathrm{st}_{\xi(\underline{\mathbb{U}},\underline{\mathbb{t}}^{(1)})}|\underline{\mathbb{T}} = \underline{\mathbb{t}}^{(1)} = \mathrm{st}_{\xi(\underline{\mathbb{U}},\underline{\mathbb{t}}^{(1)})}$$

$$\leq \mathrm{st}_{\xi(\underline{\mathbb{U}},\underline{\mathbb{t}}^{(2)})} = \mathrm{st}_{\xi(\underline{\mathbb{U}},\underline{\mathbb{t}}^{(2)})}|\underline{\mathbb{T}} = \underline{\mathbb{t}}^{(2)} = \mathrm{st}_{\xi(\underline{\mathbb{U}},\underline{\mathbb{T}})}|\underline{\mathbb{T}} = \underline{\mathbb{t}}^{(2)}.$$

Thus $\xi(U,T) \uparrow \text{ st in } T$, i.e., $f(S,T) \uparrow \text{ st in } T$.

3. Stochastically Increasing Random Variables, and Association Theorems 1.1 and 1.2 are both special cases of:

Theorem 3.1. Let T_1, \ldots, T_n be associated. Let S_1, \ldots, S_m be conditionally independent, given $T_i = t$, for all t. Let $S_i \uparrow$ st in T_i , $t = 1, \ldots, m$. Then $S_1, \ldots, S_m, T_1, \ldots, T_n$ are associated.

<u>Proof A.</u> Set $S,T = {h_1(U_1,T),...,h_m(U_m,T)},T$ in accordance with Corollary 2.2. Since $U_1,...,U_m$ are mutually independent, then $U_1,...,U_m$ are associated [E-P-W(1967), Theorem 2.1]. Since U,T are independent,

then $U_1, \ldots, U_m, T_1, \ldots, T_n$ are associated [E-P-W(1967), Property P_2]. Let $f(\underline{s},\underline{t})$, $g(\underline{s},\underline{t})$ be increasing functions such that $Cov[f(\underline{S},\underline{T}),g(\underline{S},\underline{T})]$ exists. Let $\xi(\underline{u},\underline{t}) = f[h_1(u_1,\underline{t}),\ldots,h_m(u_m,\underline{t}),\underline{t}]$, $h(\underline{u},\underline{t}) = g[h_1(u_1,\underline{t}),\ldots,h_m(u_m,\underline{t}),\underline{t}]$. ξ,η are increasing functions, and $f(\underline{S},\underline{T}), g(\underline{S},\underline{T}) = f(\underline{U},\underline{T}), g(\underline{U},\underline{T})$. Thus

$$Cov[f(S,T),g(S,T)] = Cov[\xi(U,T),n(U,T)] \ge 0,$$

and so $S_1, \ldots, S_m, T_1, \ldots, T_n$ are associated. $|\cdot|$

We find the expectation of a function f(S,T) by first conditioning on T, i.e.,

$$Ef(S,T) = E_{T}E_{S|T}f(S,T)$$

where $E_{\underline{T}}$ denotes expectation over the distribution of \underline{T} , and $E_{\underline{S}|\underline{T}}$ denotes expectation over the conditional distribution of \underline{S} , given a fixed \underline{T} .

<u>Proof B.</u> Let $f(\underline{s},\underline{t})$, $g(\underline{s},\underline{t})$ be increasing functions such that $Cov[f(\underline{S},\underline{T}),g(\underline{S},\underline{T})]$ exists. Then, dropping arguments,

(3.1)
$$\operatorname{Cov}[f,g] = \operatorname{Efg} - \operatorname{EfEg}$$

$$= \operatorname{E}_{\underline{T}}^{\operatorname{E}} \operatorname{S} | \underline{T}^{\operatorname{fg}} - \operatorname{E}_{\underline{T}}^{\operatorname{E}} \operatorname{S} | \underline{T}^{\operatorname{f}} \operatorname{E}_{\underline{T}}^{\operatorname{E}} \operatorname{S} | \underline{T}^{\operatorname{g}} \operatorname{I}$$

$$= \operatorname{E}_{\underline{T}}^{\operatorname{E}} \operatorname{S} | \underline{T}^{\operatorname{fg}} - \operatorname{E}_{\underline{T}}^{\operatorname{E}} \operatorname{S} | \underline{T}^{\operatorname{fg}} \operatorname{S} | \underline{T}^{\operatorname{g}} \operatorname{I}$$

$$+ \operatorname{E}_{\underline{T}}^{\operatorname{fg}} \operatorname{S} | \underline{T}^{\operatorname{fg}} \operatorname{S} | \underline{T}^{\operatorname{g}} \operatorname{I} - \operatorname{E}_{\underline{T}}^{\operatorname{E}} \operatorname{S} | \underline{T}^{\operatorname{f}} \operatorname{E}_{\underline{T}}^{\operatorname{g}} \operatorname{I} \underline{T}^{\operatorname{g}} \operatorname{I}$$

$$= \operatorname{E}_{\underline{T}}^{\operatorname{Cov}} \operatorname{S} | \underline{T}^{\operatorname{ff}}, g \operatorname{I} + \operatorname{Cov}_{\underline{T}}^{\operatorname{E}} \operatorname{S} | \underline{T}^{\operatorname{f}}, \underline{E}_{\underline{S}} | \underline{T}^{\operatorname{g}} \operatorname{I}.$$

Let $V_1 = {}^{st} S_1 | T = t, ..., V_m = {}^{st} S_m | T = t.$ Since $V_1, ..., V_m$ are independent, $V_1, ..., V_m$ are associated [E-P-W(1967), Theorem 2.1].

Then $\{f(S,T),g(S,T)\}|T=t=t=st\{f(S,t),g(S,t)\}|T=t=st f(V,t), g(V,t),$ and

$$Cov_{S|T=t}[f(S,T),g(S,T)] = Cov[f(V,t),g(V,t)] \stackrel{>}{=} 0,$$

by the definition of association. Thus

(3.2)
$$\mathbb{E}_{\underline{T}}^{\text{Cov}} [f(\underline{S},\underline{T}),g(\underline{S},\underline{T})] \stackrel{?}{=} 0.$$

Let $\lambda(\underline{t}) = E_{\underbrace{S|_{\underline{T}=\underline{t}}}}f(\underbrace{S},\underline{T})$, $\mu(\underline{t}) = E_{\underbrace{S|_{\underline{T}=\underline{t}}}}g(\underbrace{S},\underline{T})$. Since $f(\underbrace{S},\underline{T}) \uparrow$ st in \underline{T} by Theorem 2.3, then $\lambda(\underline{t}),\mu(\underline{t})$ are increasing functions. Since T_1,\ldots,T_n are associated,

$$(3.3) \qquad \operatorname{Cov}_{\underline{T}}[E_{\underline{S}|\underline{T}}f(\underline{S},\underline{T}),E_{\underline{S}|\underline{T}}g(\underline{S},\underline{T})] = \operatorname{Cov}[\lambda(\underline{T}),\mu(\underline{T})] \geq 0.$$

From (3.1), (3.2), and (3.3), $Cov[f(S,T),g(S,T)] \ge 0$, so that S_1,\ldots,S_m , T_1,\ldots,T_n are associated. $|\cdot|$

The following multivariate definitions of "stochastically less than" and "stochastically increasing" are of interest in the present context, and also because of their apparent relevance to reliability theory:

<u>Definition 3.2.</u> S is stochastically less than S', written $S \leq^{st} S'$, if $f(S) \leq^{st} f(S')$ for all increasing functions f(S).

<u>Definition 3.3.</u> So is stochastically increasing in T, written $S \uparrow st$ in T, if $f(S) \uparrow st$ in T for all increasing functions f(S).

It is immediate that $S \uparrow st$ in T is equivalent to $S \mid T = \frac{1}{2} \frac{(1)}{2} \frac{st}{2} S \mid T = \frac{1}{2} \frac{(2)}{2} \text{ for all } \frac{t}{2} \frac{(1)}{2} \frac{t}{2} \frac{(2)}{2}.$ From Theorem 2.3, if

 S_1, \ldots, S_m are conditionally independent, given $\underline{T} = \underline{t}$, for all \underline{t} , and S_i st in \underline{T} , $i = 1, \ldots, m$, then \underline{S} st in \underline{T} , where $\underline{S} = S_1, \ldots, S_m$.

Lemma 3.4. Let $f(\underline{s},\underline{t})$ be an increasing function. Let \underline{s} 'st in $\underline{\tau}$. Then $f(\underline{s},\underline{\tau})$ 'st in $\underline{\tau}$.

Proof. For $t^{(1)} \geq t^{(2)}$,

$$f(\underline{S},\underline{T})|\underline{T} = \underline{t}^{(1)} = \operatorname{st} f(\underline{S},\underline{t}^{(1)})|\underline{T} = \underline{t}^{(1)} \leq \operatorname{st} f(\underline{S},\underline{t}^{(2)})|\underline{T} = \underline{t}^{(1)}$$

$$\leq \operatorname{st} f(\underline{S},\underline{t}^{(2)})|\underline{T} = \underline{t}^{(2)} = \operatorname{st} f(\underline{S},\underline{T})|\underline{T} = \underline{t}^{(2)}.$$

Thus $f(\underline{S},\underline{T}) \mid \underline{T} = \underline{t}^{(1)} \leq st$ $f(\underline{S},\underline{T}) \mid \underline{T} = \underline{t}^{(2)}$ for all $\underline{t}^{(1)} \leq \underline{t}^{(2)}$, i.e., $f(\underline{S},\underline{T})$ st in \underline{T} .

 S_1, \dots, S_m are conditionally associated, given $\underline{T} = \underline{t}$, if S_1, \dots, S_m , $\underline{T} = \underline{t}$ are associated.

Theorem 3.1 is a special case of:

Theorem 3.5. Let T_1, \ldots, T_n be associated. Let S_1, \ldots, S_m be conditionally associated, given $\underline{T} = \underline{t}$, for all \underline{t} . Let \underline{S} 'st in \underline{T} .

Then $S_1, \ldots, S_m, T_1, \ldots, T_n$ are associated.

The proof follows the lines of Proof B of Theorem 3.1, using Lemma 3.4 in place of Theorem 2.3.

4. Applications and Examples

Random variables T_1, \dots, T_n are standard matrix D_n in the property of the standard standard matrix D_n is the standard matrix D_n and D_n are standard matrix D_n in the standard matrix D_n in the standard matrix D_n is the standard matrix D_n in the standard matrix D_n in the standard matrix D_n is the standard matrix D_n in the standard matrix D_n in the standard matrix D_n is the standard matrix D_n in the standard matrix D_n in the standard matrix D_n is the standard matrix D_n in the standard matrix D_n in the standard matrix D_n is the standard matrix D_n in the standard matrix D_n in the standard matrix D_n is the standard matrix D_n in the standard matrix D_n in the standard matrix D_n is the standard matrix D_n in the standard matrix D_n in the standard matrix D_n is the standard matrix D_n in the standard matrix D_n in the standard matrix D_n is the standard matrix D_n in the standard matrix D_n in the standard matrix D_n is the standard matrix D_n in the standard matrix D_n in the standard matrix D_n is the standard matrix D_n in the standard matrix D_n in the standard matrix D_n is the standard matrix D_n in the standard matrix D_n in the standard matrix D_n is the standard matrix D_n in the standard matrix D_n in the standard matrix D_n is the standard matrix D_n in the standard matrix D_n in the standard matrix D_n is the standard matrix D_n in the standard matrix D_n in the standard matrix D_n is the standard matrix D_n in the standard matrix D_n in the standard matrix D_n is the standard matrix D_n in the standard matrix D_n in the standard matrix D_n is the standard matrix D_n in the standard matrix D_n in the standard matrix D_n is the standard matrix D_n in the standard matrix D_n in the standard matrix D_n is the standard matrix D_n in the standard matrix D_n in the standard matrix D_n is the standard matrix D_n in the standard matrix D_n in the standa

if T_2 + st in T_1 , T_3 + st in $\{T_1, T_2\}, \dots, T_n$ + st in $\{T_1, \dots, T_{n-1}\}$.

Theorem 4.1. Let T_1, \dots, T_n be stochastically increasing in sequence. Then T_1, \dots, T_n are associated.

<u>Proof.</u> $\{T_1\}$ is a set of associated random variables [E-P-W(1967), Property P_3]. Since T_2 † st in T_1 , T_1 , T_2 are associated by Theorem 1.1. Continuing by induction, using Theorem 1.1, T_1 ,..., T_n are associated.

A random variable T has a wereasing failure rate (DFR) distribution [Barlow-Marshall-Proschan(1963)] if $\overline{F(t+u)}/\overline{F(t)}$ is increasing in t for all $u \ge 0$, where $\overline{F(t)} = P[T>u]$.

Example 4.2. Let $T^{(1)} \leq \ldots \leq T^{(n)}$ be the order statistics in a sample of size n from a DFR distribution. Let $D_1 = T_1$ and $D_i = T^{(i)} - T^{(i-1)}$, $i = 2, \ldots, n$. Then D_1, \ldots, D_n are stochastically increasing in sequence.

Proof. Note that

$$P[D_{i+1} | u | D_{i} = d_{1}, \dots, D_{i} = d_{i}] = \left\{ \frac{\overline{F}(d_{1} + \dots + d_{i} + u)}{\overline{F}(d_{1} + \dots + d_{i})} \right\}^{n-i}$$

is increasing in d_1, \ldots, d_i for all $u \ge 0$. Thus $D_{i+1} \uparrow$ st in D_1, \ldots, D_i , and so D_1, \ldots, D_n are stochastically increasing in sequence. $|\cdot|$

It follows from Theorem 4.1 that D_1, \dots, D_n are associated.

A stochastic process $\{X(t), t \in \tau\}$ is associated in time if the random variables $X(t_1), \ldots, X(t_k)$ are associated for all k and all $\{t_1, \ldots, t_k\} \in \mathbb{R}$. $\{X(t), t \in \tau\}$ is stackastically increasing in sequence for all k and if $X(t_1), \ldots, X(t_k)$ are stochastically increasing in sequence for all k and

all $\{t_1 < \ldots < t_k\} \subset \tau$. It is equivalent to say that $\{X(t), t \in \tau\}$ is stochastically increasing in time if $X(t_k) \uparrow st$ in $\{X(t_1), \ldots, X(t_{k-1})\}$ for all $\{t_1 < \ldots < t_k\} \subset \tau$.

Theorem 4.3. Let $\{X(t), t \in \tau\}$ be stochastically increasing in time. Then $\{X(t), t \in \tau\}$ is associated in time.

<u>Proof.</u> For any k and $\{t_1 < ... < t_k\} \subset \tau$, $X(t_1)$, ..., $X(t_k)$ are stochastically increasing in sequence. By Theorem 4.1 $X(t_1)$, ..., $X(t_k)$ are associated. Thus $\{X(t), t \in \tau\}$ is associated in time.

Theorem 4.4. Let $\{X(t), t \in \tau\}$ be a Markov process. Let $X(t_2) \uparrow$ st in $X(t_1)$ for all $\{t_1 < t_2\} \subset \tau$. Then $\{X(t), t \in \tau\}$ is stochastically increasing in time.

 $\begin{array}{lll} & \underline{\text{Proof.}} & \text{Let } \big\{ t_1 < \ldots < t_k \big\} \subset \tau. & \text{Since } \big\{ \textbf{X}(\textbf{t}), \ \textbf{t} \in \tau \big\} & \text{is a Markov} \\ \\ & \text{process, } & \textbf{X}(\textbf{t}_k) \, \big| \big\{ \textbf{X}(\textbf{t}_1) = \textbf{x}_1, \ \ldots, \ \textbf{X}(\textbf{t}_{k-1}) = \textbf{x}_{k-1} \big\} = \overset{\text{st}}{} & \textbf{X}(\textbf{t}_k) \, \big| \, \textbf{X}(\textbf{t}_{k-1}) = \textbf{x}_{k-1}, \\ \\ & \text{all } & \textbf{x}_1, \ \ldots, \ \textbf{x}_{k-1}. & \text{Let } & \textbf{x}_1 \leq \textbf{y}_1, \ \ldots, \ \textbf{x}_{k-1} \leq \textbf{y}_{k-1}. & \text{Since } & \textbf{X}(\textbf{t}_k) + \text{st} \\ \\ & \text{in } & \textbf{X}(\textbf{t}_{k-1}), \ \textbf{X}(\textbf{t}_k) \, \big| \, \textbf{X}(\textbf{t}_{k-1}) = \textbf{x}_{k-1} \leq \overset{\text{st}}{} & \textbf{X}(\textbf{t}_k) \, \big| \, \textbf{X}(\textbf{t}_{k-1}) = \textbf{y}_{k-1}. & \text{Then} \\ \end{array}$

$$\begin{aligned} & \mathbf{X}(\mathbf{t}_{k}) \mid \left\{ \mathbf{X}(\mathbf{t}_{1}) = \mathbf{x}_{1}, \dots, \mathbf{X}(\mathbf{t}_{k-1}) = \mathbf{x}_{k-1} \right\} = \mathbf{st} \ \mathbf{X}(\mathbf{t}_{k}) \mid \mathbf{X}(\mathbf{t}_{k-1}) = \mathbf{x}_{k-1} \\ & \leq \mathbf{st} \ \mathbf{X}(\mathbf{t}_{k}) \mid \mathbf{X}(\mathbf{t}_{k-1}) = \mathbf{y}_{k-1} = \mathbf{st} \ \mathbf{X}(\mathbf{t}_{k}) \mid \left\{ \mathbf{X}(\mathbf{t}_{1}) = \mathbf{y}_{1}, \dots, \mathbf{X}(\mathbf{t}_{k-1}) = \mathbf{y}_{k-1} \right\}. \end{aligned}$$

Thus $X(t_k)$ ist in $\{X(t_1), \dots, X(t_{k-1})\}$, i.e., $\{X(t), t \in \tau\}$ is stochastically increasing in time. $|\cdot|$

Example 4.5. Let $\{X(t), t \in \tau\}$ be a Markov process such that X(t) = 0 or 1 for each $t \in \tau$. Let

$$P[X(t_2) = 1 \mid X(t_1) = 1] = p(t_1, t_2)$$

 $P[X(t_2) = 0 \mid X(t_1) = 0] = q(t_1, t_2)$

where $p(t_1,t_2)+q(t_1,t_2)\geq 1$ for all $\{t_1 < t_2\} \subset \tau$. Then $\{X(t), t \in \tau\}$ is stochastically increasing in time.

Proof. For $\{t_1 < t_2\} \subset \tau$, set $U(t_1, t_2) = {}^{st} X(t_2) | X(t_1) = 1$, $V(t_1, t_2) = {}^{st} X(t_2) | X(t_1) = 0$. Then $P[V(t_1, t_2) = 1] = 1 - q(t_1, t_2)$ $\leq p(t_1, t_2) = P[U(t_1, t_2) = 1]$, i.e., $V(t_1, t_2) \leq {}^{st} U(t_1, t_2)$. Thus $X(t_2) \uparrow {}^{st}$ in $X(t_1)$, and by Theorem 4.4 $\{X(t), t \in \tau\}$ is stochastically increasing in time. $|\cdot|$

In reliability theory, processes of the type considered in Example 4.5 are basic models for the performance of a device subject to alternate failure and repair, where X(t)=1 if the device is functioning at time t, X(t)=0 if the device is failed at time t. If $\tau=\left\{0,1,\ldots\right\}$ and p(k,k+1)=p, q(k,k+1)=q, then $\left\{X(t),\,t\in\tau\right\}$ corresponds to an alternating renewal process where time from repair to failure has a geometric distribution with parameter p, and time from failure to repair has a geometric distribution with parameter q. This geometric-geometric performance process is stochastically increasing in time if $p+q\geq 1$. If $\tau=\left\{0,+\infty\right\}$ and

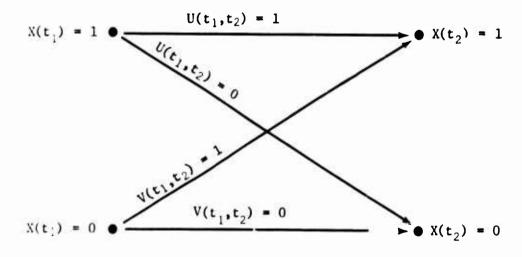
$$p(t_1, t_2) = (\lambda + \mu)^{-1} \{ \mu + \lambda \exp[-(\lambda + \mu)(t_1 - t_1)] \}$$

$$q(t_1, t_2) = (\lambda + \mu)^{-1} \{ \lambda + \mu \exp[-(\lambda + \mu)(t_2 - t_1)] \},$$

then $\{X(t), t \in \tau\}$ corresponds to an alternating renewal process where

time from repair to failure has an exponential distribution with parameter , and time from failure to repair has an exponential distribution with parameter μ . Since $\mu(t_1,t_2)+\mu(t_1,t_2)\geq 1$ for all $\mu_1<\mu_2<\mu_3> 0$, the exponential-exponential performance process is stochastically increasing in time. It follows from Theorem 4.3 that the exponential-exponential performance process and the geometric-geometric performance process with $\mu_1+\mu_2>0$ are associated in time [cf. E-P-W(1966)]. Some properties and applications in reliability theory of performance processes that are associated in time are discussed in E-P-W(1966).

In the process of Example 4.5 $\{X(t_1), X(t_2)\}$ are, for $\{t_1 < t_2\} \subset \tau$, stochastically representable by $X(t_1)$ and transition random variables $U(t_1,t_2) = {}^{\text{st}} X(t_2) | X(t_1) = 1, \quad V(t_1,t_2) = {}^{\text{st}} X(t_2) | X(t_1) = 0, \quad \text{such}$ that $X(t_1), U(t_1,t_2), V(t_1,t_2)$ are mutually independent.



 $\begin{array}{l} \{X(t_1),\ X(t_2)\} = \begin{array}{l} \text{st} \ \{X(t_1),\ X(t_1) \, \mathbb{U}(t_1,t_2) \, + \, [1-X(t_1)] \, \mathbb{V}(t_1,t_2)\}, \quad \text{and so} \\ X(t_1) \cdot \text{st} \quad \text{in} \quad X(t_1) \quad \text{is equivalent to} \quad \mathbb{V}(t_1,t_2) \, \leq \begin{array}{l} \text{st} \ \mathbb{U}(t_1,t_2). \quad \text{Setting} \\ \mathbb{U}(t_1,t_1),\ \mathbb{V}(t_1,t_2) \quad \text{independent is convenient, but not essential to the} \\ \text{representation.} \end{array}$

In a frequently studied reliability model involving a complex of n identical devices, the functioning devices are in various degrees of service or standby for service, and the failed devices are in various degrees of repair or standby for repair. In this model the basic descriptor of performance is X(t) = the number of devices functioning at time t. The following generalization of Example 4.5 covers a variety of cases in which the process $\{X(t), t \in t\}$ is stochastically increasing in time, and thus associated in time.

Example 4.6. Let $\{X(t), t \in \tau\}$ be a Markov process such that X(t) = 0 or 1 or ... or n for each $t \in \tau$. For all $\{t_1 < t_2\} \subset \tau$ let

$$\begin{split} & X(t_1) \mid X(t_1) = i = \frac{st}{U_1}(t_1,t_2) + \ldots + U_i(t_1,t_2) + V_{i+1}(t_1,t_2) + \ldots + V_n(t_1,t_2), \\ & i = 0,\ldots,n, \quad \text{where} \quad U_i(t_1,t_2) = 0 \text{ or } 1, \quad V_i(t_1,t_2) = 0 \text{ or } 1, \quad i = 1,\ldots,n, \\ & \{U_1(t_1,t_1), V_1(t_1,t_2)\}, \quad \ldots, \quad \{U_n(t_1,t_2), V_n(t_1,t_2)\} \quad \text{are mutually independent couples, and} \quad V_i(t_1,t_2) \leq \frac{st}{U_i(t_1,t_2)}, \quad i = 1,\ldots,n. \quad \text{Then} \\ & \{X(t), \ t \in \tau\} \quad \text{is stochastically increasing in time.} \end{aligned}$$

Proof. Set

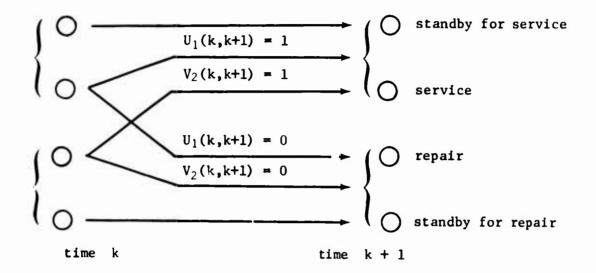
$$Z_{\mathbf{j}}(\mathbf{t}_{1}) \mid X(\mathbf{t}_{1}) = \mathbf{i} = \mathbf{st} \begin{cases} V_{\mathbf{j}}(\mathbf{t}_{1}, \mathbf{t}_{2}) & \text{if } \mathbf{i} \subseteq \mathbf{j} \\ V_{\mathbf{j}}(\mathbf{t}_{1}, \mathbf{t}_{1}) & \text{if } \mathbf{i} \leq \mathbf{j}, \end{cases}$$

j = 1,...,n. Since $V_j(t_1,t_1) = \frac{st}{t_1}U_j(t_1,t_1)$, $Z_j(t_1) + st$ in $X(t_1)$. Let $Z_1(t_1),...,Z_n(t_n)$ be conditionally independent, given $X(t_1) = i$, i = 0,...,n. Then

We illustrate the application of Example 4.6 for plans involving two identical devices, n=2, where time is measured in discrete cycles, say $i=\{0,1,\ldots\}$. We suppose that devices fail or are repaired within cycles, and that devices are transferred from standby for service to service, service to standby for repair, etc., between cycles.

We view the experience of a device within each cycle as independent of its experience in preceding cycles, and dependent only on the type of service or repair it is subject to on that cycle.

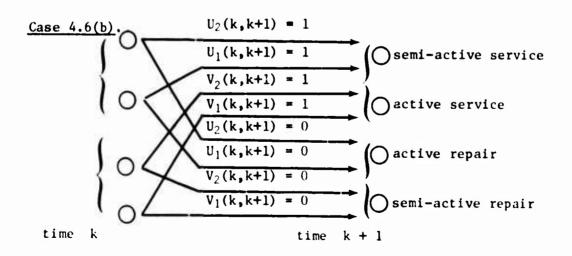
Case 4.6(a).



Assuming $X(t_k)$, $U_1(k,k+1)$, $V_2(k,k+1)$ mutually independent:

$$X(k+1) | X(k) = 2 =$$
st $U_1(k,k+1) + 1$.
 $X(k+1) | X(k) = 1 =$ st $U_1(k,k+1) + V_2(k,k+1)$.
 $X(k+1) | X(k) = 0 =$ st $0 + V_2(k,k+1)$.

X(k+1) † st in X(k) by Example 4.6.



Assuming X(k), $U_1(k,k+1)$, $U_1(k,k+1)$, $V_1(k,k+1)$, $V_1(k,k+1)$ mutually

independent:

By Example 4.6,
$$X(k+1) \uparrow st$$
 in $X(k)$ if $V_1(k,k+1) \le {}^{st} U_1(k,k+1)$ and $V_2(k,k+1) \le {}^{st} U_2(k,k+1)$, e.g., if $U_1(k,k+1) \le {}^{st} U_2(k,k+1)$, $V_1(k,k+1) \le {}^{st} V_2(k,k+1)$, and $V_2(k,k+1) \le {}^{st} U_1(k,k+1)$.

Further applications of Example 4.6, e.g., to cases in which time is measured continuously, will be considered in a forthcoming document.

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